

### Overview of the LSST Data Management Activities at LLNL

Ghaleb Abdulla

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## Overview of the LSST Data Management Activities at LLNL

#### Ghaleb Abdulla

**Center for Applied Scientific Computing** 

**Lawrence Livermore National Lab** 



#### **Overview**

- Project, Team, collaborators
- Modular data-centric pipeline
- Support for real-time query processing
- H/W and S/W HPC architectures
- Other research topics
- Partnering approaches



Work in progress



#### **LLNL/LSST Data Management**

- REal-time QUEry of Sensor sTreaming Data (REQUEST)
  - —http://www.llnl.gov/CASC/request/
- Team
  - —Ghaleb Abdulla (100%,PL)
  - —Kem Cook (20%, Project Scientist)
  - —Sergei Nikolaev (50%, Application Scientist)
  - Marcus Miller (100%, System Architect)
  - —Jim Garlick (50%, Advanced Developer)
  - —Postdoc (100%, database/data-streaming-TBD)
- Collaborators
  - —Berkeley
  - -NCSA



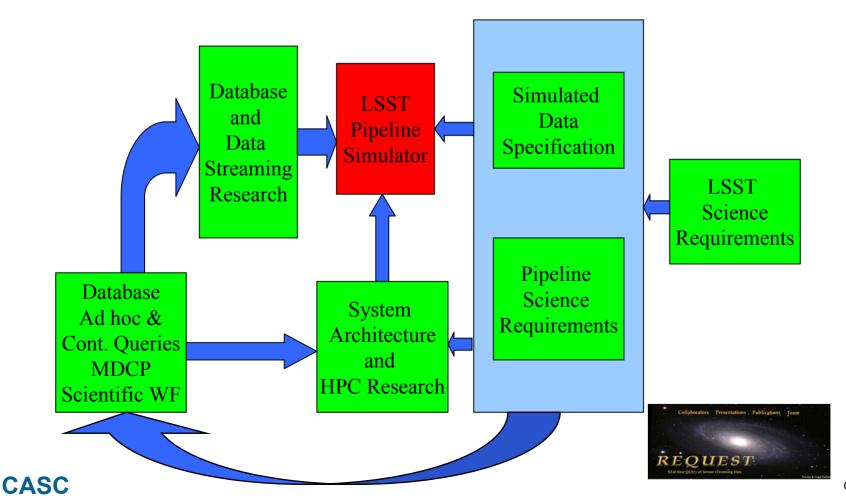
#### **LLNL Strengths**

- Big cluster machines (MCR, Thunder, BlueGene/L, ....)
- HPC system architecture/development expertise
- Data-management research
- Opportunities to interact with the application scientists
- Strong potential collaborators (Berkeley, NCSA, ...)

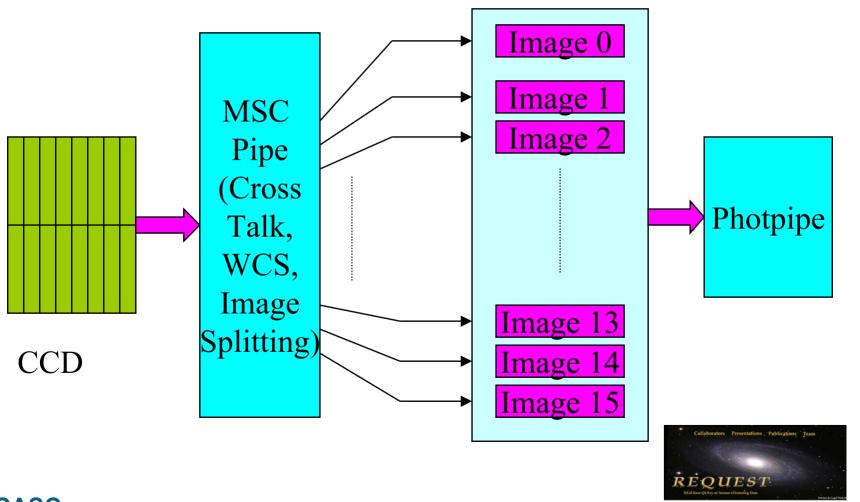


#### **Overview Of LLNL Research Activities**

The objective is to build an LSST data pipeline simulator



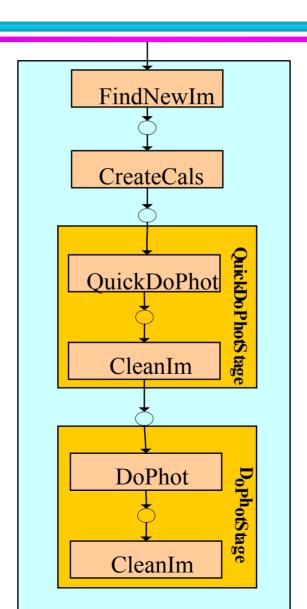
#### SuperMacho Data Analysis Stages



**CASC** 

GA 6

## **Photpipe**





# LLNL prototyping strategy and plans

- Use SM data pipeline to:
  - —Test scalability on parallel cluster architectures
  - —Research support of real time analysis
  - Generate synthetic data to test with anticipated LSST data rates
- Research thrusts to achieve these plans:
  - Modular data-centric pipeline
  - Data streaming and support of real-time continuous queries
  - —H/W and S/W HPC architectures



#### Modular data-centric pipeline

- Collaborated with the database group at Berkeley (GridDB)
- Modeled photpipe, a super macho data pipeline, as one big box; implementation using GridDB is underway
- Started looking into modeling the individual pipeline components.



#### Milestones (9/2004)

- Show the utility of using a modular data centric software architectures by implementing the super macho data pipeline under GridDB
- Feed the output from the analysis into a database system
- Test the implementation with example queries



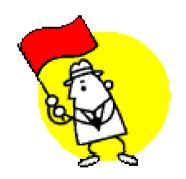
# Real time query and data analysis support

- Identify where real-time query support is needed within the data pipelines (Work in progress)
- Investigate using the data streaming system ,Telegraph, from Berkeley to support the real time query processing



### Milestones (9/2004)

- Identify a set of real-time continuous queries
- Investigate real-time support of the pre-defined set of queries for alerting services
- Investigate methods to integrate "Telegraph" with the the data pipeline work





#### **High Performance Computing**

- Install and run the data pipelines on a cluster computer
- Characterize the data pipeline performance
- Investigate supporting scientific work-flows in a distributed computing environment



#### Milestones (9/2004)

- Port the existing Super Macho (SM) astrophysics image analysis pipeline to a large distributed memory, multi-node cluster platform employing a global parallel file system
- Evaluate the performance of the pipeline
- Profile the code to determine which sections yield the highest compute and I/O costs
- Characterize the common communication and I/O access patterns.



#### Milestones (9/2005)

- Generic implementation of GridDB that allows creating new analysis data pipelines with new functionalities by swapping algorithms and support parallel workflow scheduling on massively parallel clusters
- GUI to support query capabilities and easy workflow construction
- Integrate a streaming system to support real-time continuous queries with GridDB
  - Scalability issues
- Start tests of an LSST prototype pipeline



#### Other R&D issues

- LSST Digital Library
  - —Real-time queries on digital libraries
  - Massive updates
  - Indexing issues (incremental, rebuilding the indexes)
- Confidence level and relevance in the final science results



## Partnering Approaches What We Have to Offer

- Technical contribution
  - Provide a framework for building new modular data centric analysis pipelines.
  - Help define requirements for flexible system architectures needed to support LSST science applications
  - Solve the technical research challenges that face the DM LSST efforts
- Informal leadership
  - —Help setting up the direction of LSST DM efforts
  - —Contribute in the grant review process
  - —Help in other areas?



## Partnering Approaches Our Needs

- Money?
  - Enable academic collaborations
- Collaboration
  - —Help us organize meetings and workshops
  - —Provide us with feedback
  - Make our work known to other groups
- Include us in your plans



### **Questions**

